

The ACA Medicaid Rebate Rule Change: Impact on Pricing and Innovation

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Motivation I: How do firms respond to government pricing & subsidies?

- ▶ Price regulation/subsidies are common in welfare programs
 - ▶ Medicare/Medicaid reimbursement
 - ▶ Premium subsidies for insurance plans
 - ▶ Rent control/public housing vouchers
 - ▶ Federal financial aid for colleges
 - ▶ Food stamps

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- ▶ Regulation leads to distortions
 - ▶ Government benchmarks can anchor private prices (e.g. Medicare reimbursement rates)
 - ▶ Fluctuating benchmarks that are tied to equilibrium prices will change firm strategic incentives

Motivation II: the Medicaid Drug Rebate Program

Duggan and Scott-Morton (2006): drugs with high Medicaid Market Share (MMS)

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- ▶ introduce more line extensions

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- ▶ Cross-sectional evidence on list prices only
- ▶ MDRP contains provisions that DSM06 doesn't consider

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However:

- ▶ Cross-sectional evidence on list prices only
- ▶ MDRP contains provisions that DSM06 doesn't consider
- ▶ Medicaid has changed since 2006:
 - ▶ Medicare Part D covers dual eligibles
 - ▶ MDRP formula increased minimum rebate in 2010
 - ▶ New data on estimated net prices is available

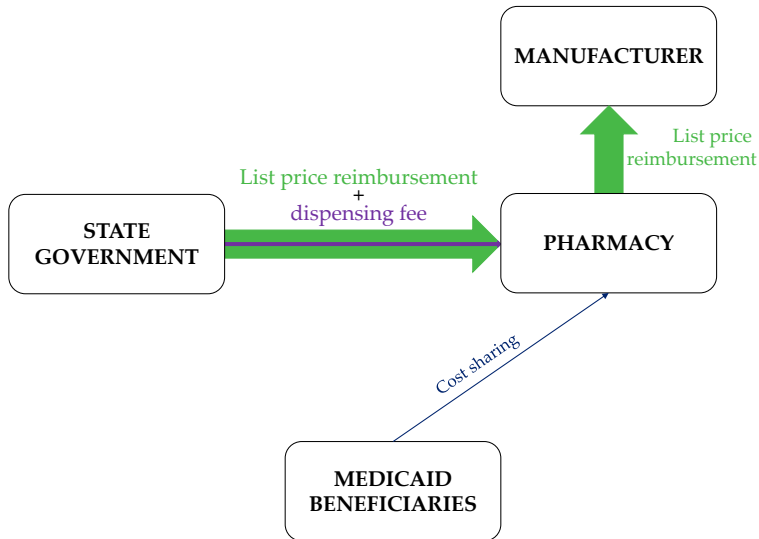
We find that the MDRP has a more nuanced impact than previously thought

Main takeaways

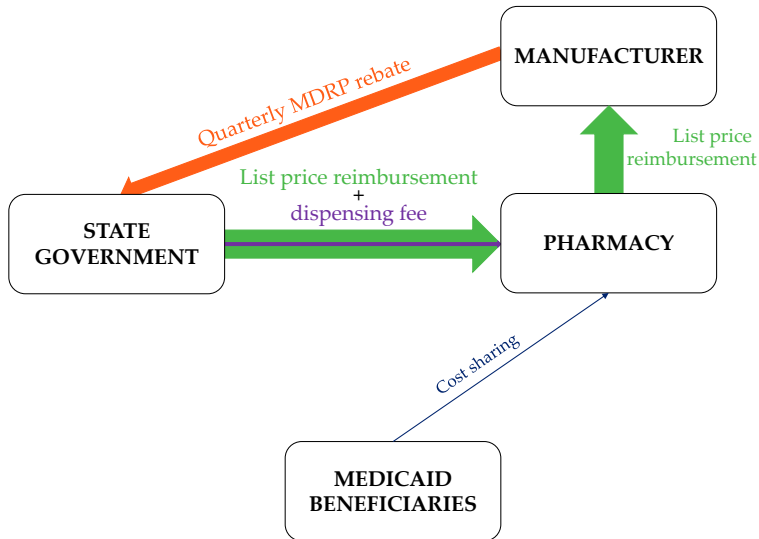
- ▶ Drugs with high Medicaid exposure:
 - ▶ Increase prices at a lower rate
 - ▶ Give lower commercial rebates
 - ▶ Launch line extensions at a higher rate
- ▶ (Not today): Little evidence of higher launch drug prices
- ▶ 2010 increase in minimum rebate enhanced positive/reduced negative effects

Overview of the Medicaid Drug Rebate Program

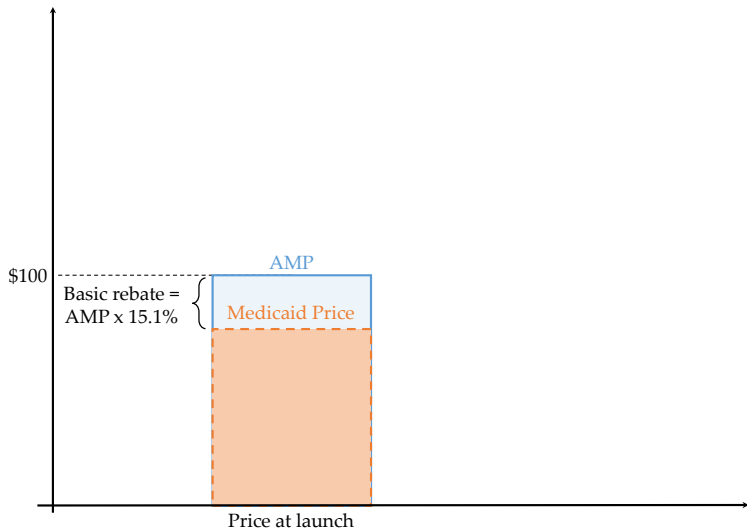
Medicaid initially pays for drugs at list prices



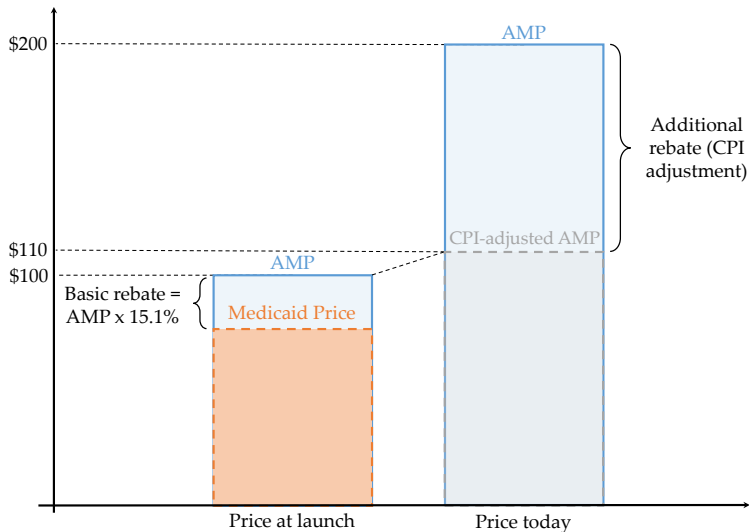
Manufacturers then send lump-sum rebates



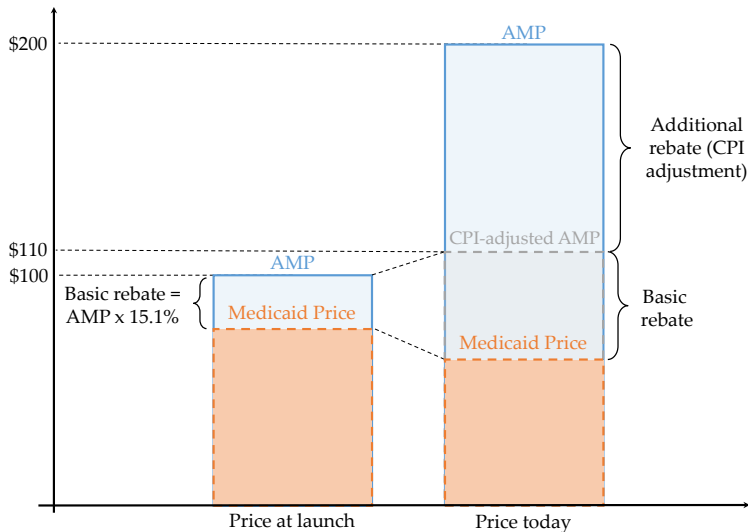
Quarterly rebate is a fraction of list price



Formula implies the price is anchored to launch price



And that prices fall if price growth exceeds inflation



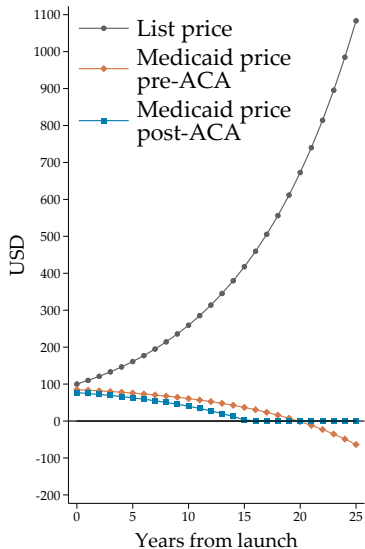
There are a few other relevant features of the program

- ▶ Medicaid is entitled to the “best price” if it is lower than $AMP - 15.1\%$
 - ▶ Difference between AMP and best price becomes new rebate %
 - ▶ Basically a “most-favored nation” clause
- ▶ Line extensions “reset the clock” on price
 - ▶ Line extensions are versions of the drug with the same active ingredient but different form/strength
 - ▶ Firms get to set a new initial price for line extensions

CMS changed the formula starting in January 2010

Two main changes:

- ▶ Base rebate increased from **15.1%** to **23.1%** of AMP
- ▶ Max rebate capped at **100%**



Optimal firm behavior

Setting

- ▶ Increasing demand
- ▶ Medicaid demand inelastic
- ▶ Maximum initial price bounded
- ▶ Firm sets AMP, discount
- ▶ Medicaid formula:

$$p_t^{\text{Med}} = \min(p_0, p_t) - p_t \times \max(r, d_t)$$

where

- ▶ p_0 is the launch AMP
- ▶ p_t is the AMP in period t
- ▶ d_t is the discount granted to commercial payer

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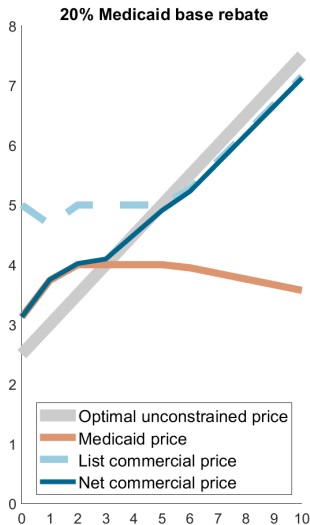
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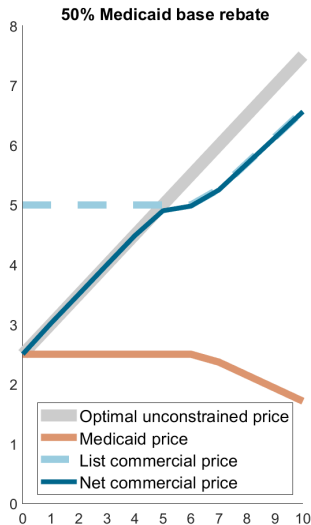
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From the model we derive a few testable hypotheses

Drugs with high Medicaid exposure will have:

1. Slower list price growth (but possibly higher launch price)
2. Lower discounts to commercial payers (to avoid triggering the best-price clause)
3. Higher probability of introduction of line extensions

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After the formula changes:

1. Even slower list price growth (unless discount is at 100%)
2. Higher-than-before discounts to commercial payers
3. Even higher probability of line extensions

Data and Empirical Strategy

Data

- ▶ SSR Health (~1,000 brand drugs, quarterly from 2007-2019)
 - ▶ Gross sales, volume
 - ▶ Net sales (obtained from SEC filings)
 - ▶ Product name level (e.g. ABILIFY)
- ▶ Medicaid PUF (quarterly from 1990-2019)
 - ▶ Gross sales, volume
 - ▶ NDC level (product-form-strength), e.g. ABILIFY-TABLET-20MG

Key variables

- ▶ **Medicaid Market Share:**

$$MMS = \frac{\text{Medicaid sales}}{\text{Invoice sales}}$$

- ▶ **List price:** measured as Wholesale Acquisition Cost (WAC)
- ▶ **non-Medicaid discount:** estimated as

$$1 - \frac{\text{Net sales} - \text{net Medicaid sales}}{\text{Invoice sales} - \text{Medicaid sales}}$$

- ▶ **Number of line extensions:** new NDC with new form or strength

Issue: invoice sales are underreported for many drugs



- ▶ Problem for many specialty, physician-administered drugs
- ▶ Example: Eylea (Aflibercept, macular degeneration)
 - ▶ WAC sales (2013): ~6 million
 - ▶ Medicaid sales: ~5 million

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- ▶ Problem for many specialty, physician-administered drugs
- ▶ Example: Eylea (Aflibercept, macular degeneration)
 - ▶ WAC sales (2013): ~6 million
 - ▶ Medicaid sales: ~5 million
 - ▶ Net sales: ~1.5 billion
- ▶ Solution: drop drugs with net sales > invoice sales over the life-cycle

Estimation exploits variation in exposure to Medicaid to estimate diff-in-diff around 2010

- ▶ Two independent variables of interest
 1. Medicaid Market Share → matters for price
 2. Medicaid sales → matters for line extensions
- ▶ Sample: drugs launched in 2007 w/ positive sales in 2009
- ▶ Two regression designs:
 1. Linear interaction of MMS/Sales with policy change
 2. Interaction of quartiles of MMS/sales with policy change

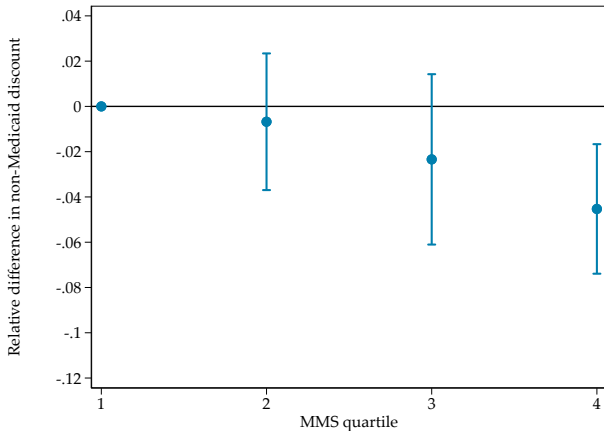
Results I: Price Distortions

List price of drugs with high MMS grows more slowly

$$\log(WAC_{it}) = \alpha_i + \delta_t + \beta_1 \times MMS_i \times (t - 2007)$$

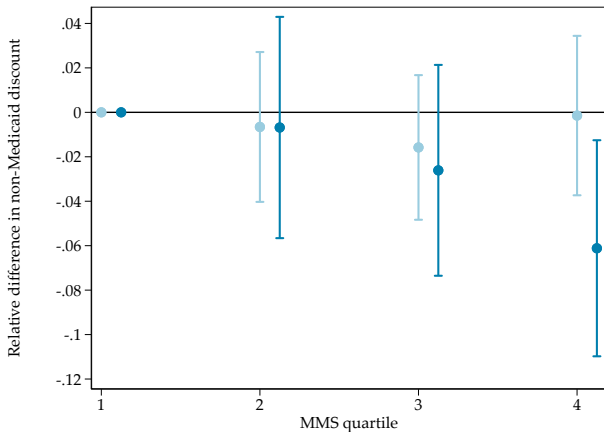
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and even more slowly after 2010

$$\log(WAC_{it}) = \alpha_i + \delta_t + \beta_1 \times MMS_i \times (t - 2007) \\ + \beta_2 \times MMS_i \times PostACA_t \times (t - 2010)$$

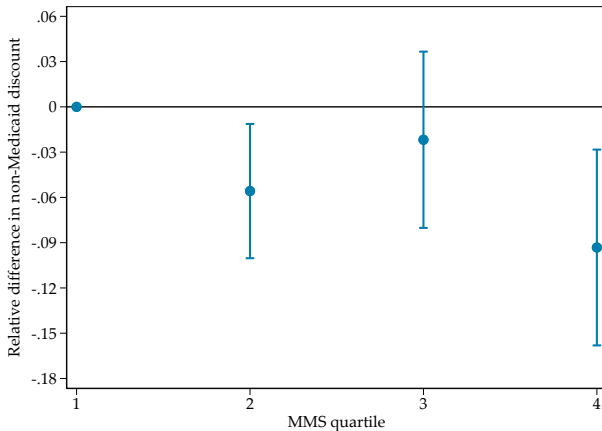


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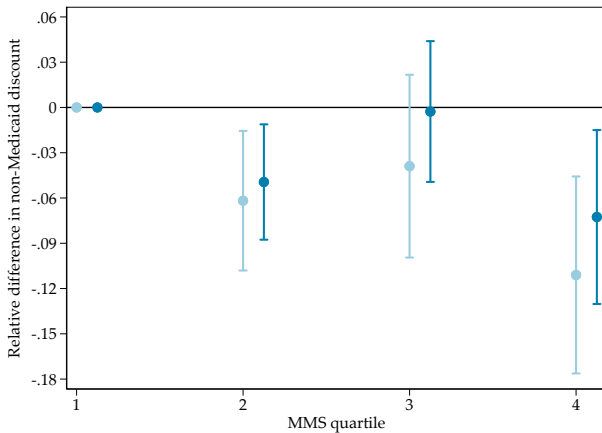
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$$\text{Discount}_{it} = \delta_t + \beta_1 \times \text{MMS}_i$$



but less so after the change in the formula

$$Discount_{it} = \delta_t + \beta_1 \times MMS_i + \beta_2 \times MMS_i \times PostACA_t$$



Results II: Innovation Distortions

Line extensions are more likely for drugs with high Medicaid sales

$$\lambda(\text{age} | X) = \lambda_0(\text{age}) \times \exp(X)$$

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	All Line Extensions
High Med Sales	2.278**
	(0.367)
Post-ACA	
High Med Sales × Post-ACA	
<i>N</i>	552

and even more so after the ACA

$$\lambda(\text{age} | X) = \lambda_0(\text{age}) \times \exp(X)$$

	All Line Extensions	
High Med Sales	2.278** (0.367)	1.725** (0.364)
Post-ACA		0.947 (0.246)
High Med Sales × Post-ACA		2.078* (0.691)
<i>N</i>	552	552

Which line extensions are more profitable under Medicaid rules?

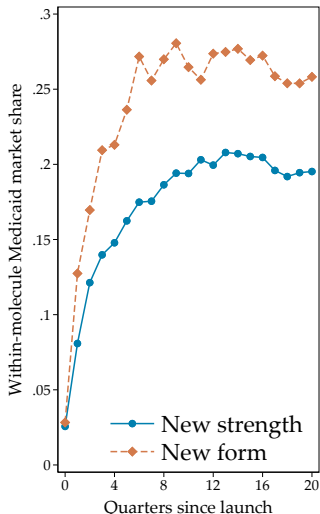
Intuition

- ▶ Key of a line extension is to get people to switch
- ▶ Higher quality line extensions can get more people to switch
- ▶ Marginal gain from extra switchers increases with base rebate

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Firms develop higher-quality line extensions post-ACA

$$\lambda(\text{age} | X) = \lambda_0(\text{age}) \times \exp(X)$$

	New Form		New Strength	
High Med Sales	2.459** (0.600)	1.439 (0.515)	2.370** (0.404)	1.939** (0.433)
Post-ACA		0.937 (0.348)		1.110 (0.304)
High Med Sales × Post-ACA		3.029* (1.500)		1.697 (0.589)
<i>N</i>	552	552	552	552

Conclusion

The Medicaid Drug Rebate Program affects pricing and R&D strategies of firms

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Predicting effect of policy change is not easy

- ▶ Simple prediction: \uparrow mandatory rebate \implies \uparrow distortion
- ▶ But firms face a lot of constraints that are hard to model
- ▶ These constraints affect predictions
- ▶ In this case, the reform turns out to be relatively benign